

Old. Likelihood in the paper.

$$\log L(a, v, k) = \sum_{x,t} \left\{ D_{x,t} (a_x + v_x k) - E_{x,t}^c \exp(a_x + v_x k) \right\} + \text{constant}$$

New Likelihood function with  $\beta_x$  parameter.

$$\log L(\beta, v, k) = \sum_{x,t} \left\{ D_{x,t} (\beta_x \cdot \log e_{\theta,t} + v_x k) - E_{x,t} \cdot \exp(\beta_x \cdot \log e_{\theta,t} + v_x k) \right\} + c.$$

Taking the first and second partial derivatives with respect to  $\beta_x$ , leads to

$$\begin{aligned} \frac{\partial L}{\partial \beta_x} &= \sum_t \left\{ D_{x,t} \cdot \log e_{\theta,t} - E_{x,t} \cdot \exp(\beta_x \log e_{\theta,t} + v_x k) \cdot \log e_{\theta,t} \right\} \\ &= \log e_{\theta,t} \left[ \sum_t (D_{x,t} - \hat{D}_{x,t}) \right] \end{aligned}$$

$$\begin{aligned} \frac{\partial^2 L}{\partial^2 \beta_x} &= \log e_{\theta,t} \left[ \sum_t (-\hat{D}_{x,t}) \right] \cdot \log e_{\theta,t} \\ &= (\log e_{\theta,t})^2 \left[ -\sum_t (\hat{D}_{x,t}) \right] \end{aligned}$$

Using this and the updating scheme proposed by Brauhns et al. given by

$$\hat{Q}(v+1) = \hat{Q}(v) - \frac{\partial L(v) / \partial \theta}{\partial^2 L(v) / \partial^2 \theta} ,$$

The parameter  $\beta_x$  can be updated as

$$\hat{\beta}_x(w+1) = \hat{\beta}_x(w) - \frac{\sum_t (D_{x,t} - \hat{D}_{x,t})}{-\sum_t \hat{D}_{x,t} \cdot \log \hat{e}_{o,t}} .$$